

# Where to Look? Predictive Perception with Application to Planetary Exploration

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#### Introduction

#### Current



- Four rovers on Mars
- Looking for the evidence of life
  - Frequent stops for scientific interests
- Day-to-day operations
  - Planning tactical activities for 1-3 sols (Martian day) in a single ground-inthe-loop cycle
  - "Restricted sols" due to Earth-Mars time gap
- Limited use of AutoNav (autonomous navigation)

#### Future (2020 and beyond)

Mars 2020

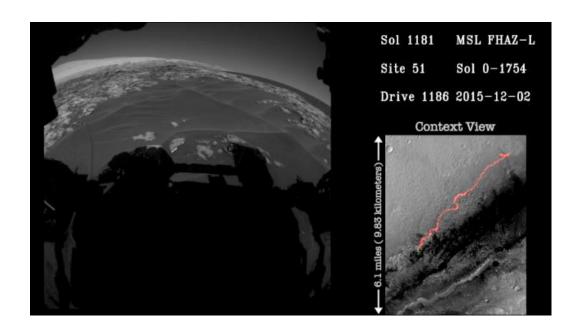
Mars Sample Return Fetch rover (concept)

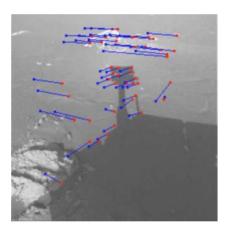




- More drive-oriented missions
  - e.g., Collect sample tubes and return to the base before the launch chance
  - Travel far longer distance per sol
- Higher dedication to AutoNav
  - Drive beyond the line of sights
  - Drive on restricted sols
- Increased onboard resources
  - Vision-dedicated processors (Mars 2020) and multi-core general-purpose processors (future missions)

## **Computer Vision in Autonomous Navigation**

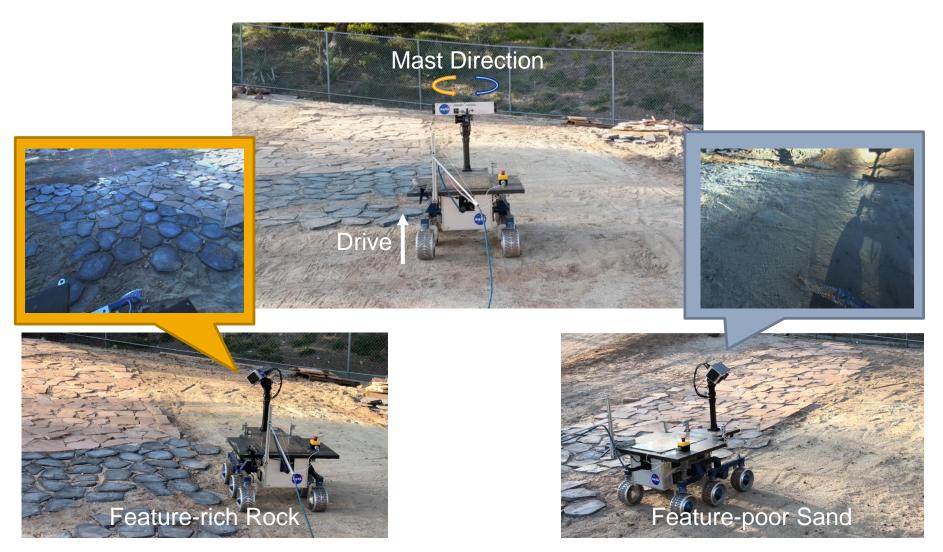




- Computer vision is a major source for autonomous navigation
- Today's vision system requires human intervention (e.g., manual mast pointing) to deal with challenges such as
  - Texture-poor terrain: Lack of salient features
  - Self-shadow: Confusion in visual feature tracking

## **Perception-aware Motion Planning**

What is the best actions to maximize perception performance?

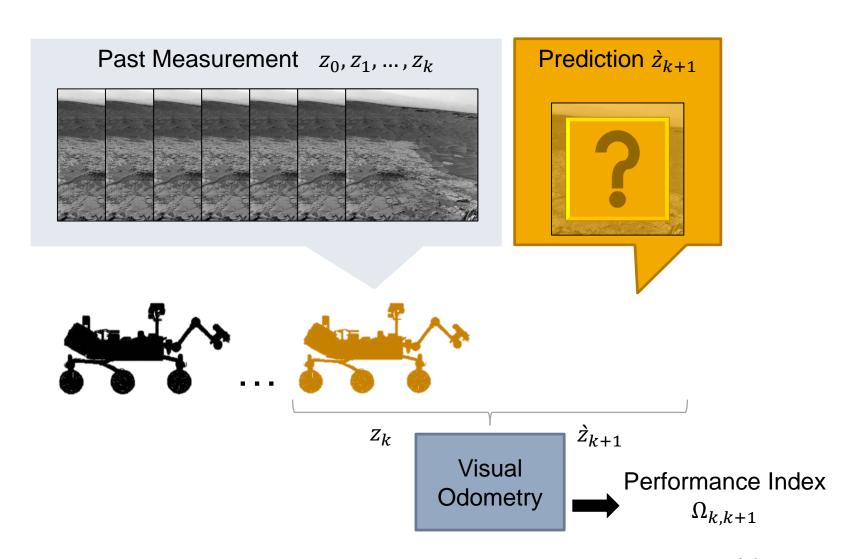


## **Key Technologies**

- 1. Performance prediction with future measurements
  - Most-likely measurement prediction with stochastic map
- 2. Optimal mast trajectory planning
  - Spatio-temporal RRT\*

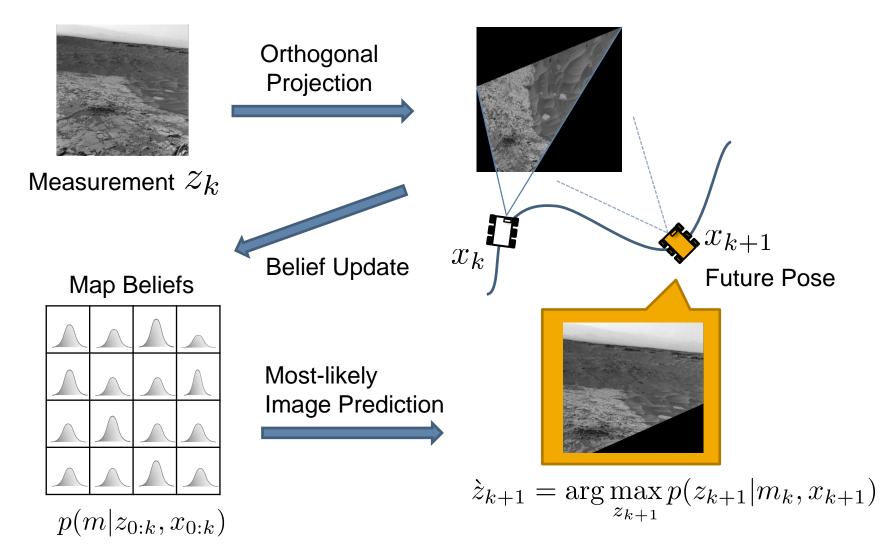
#### 1. Performance Prediction

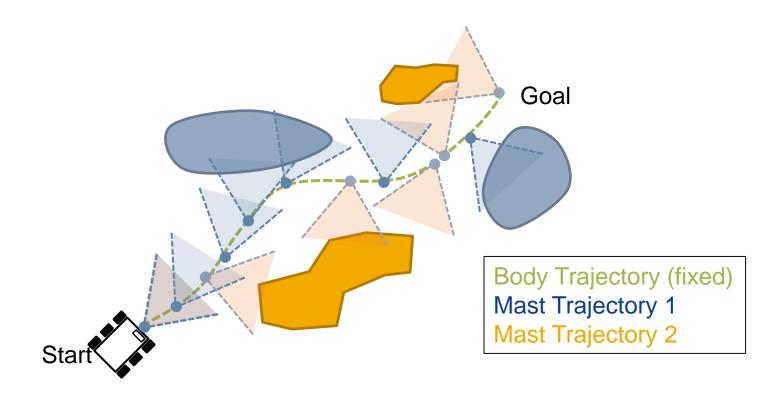
Quantize the performance of future measurements



#### 1. Performance Prediction

How to predict most-likely future measurement?



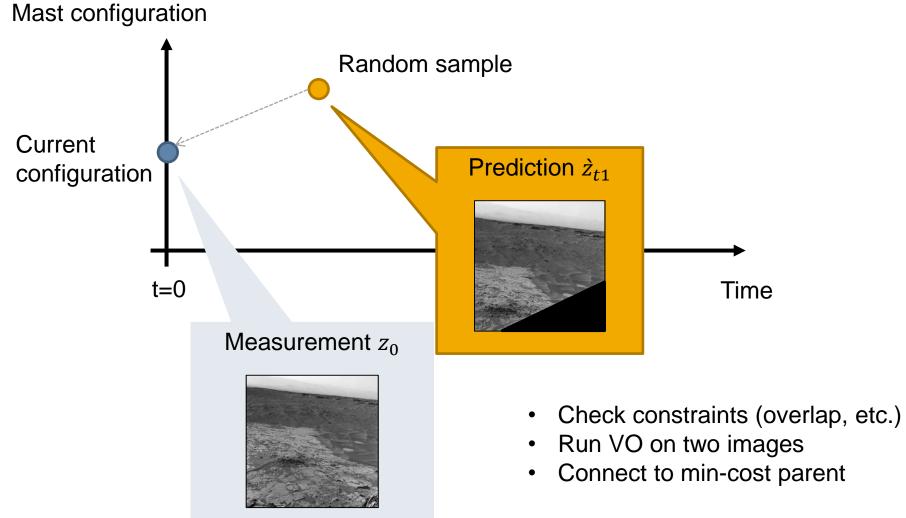


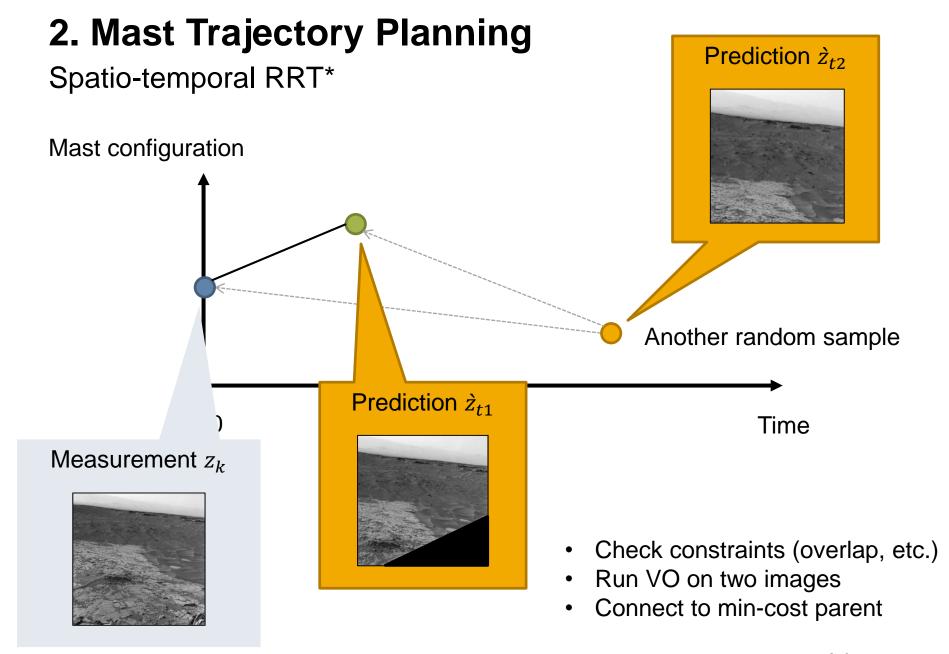
Mast state: (Pan, Tilt, Time)

Where to point mast?

When to take image?

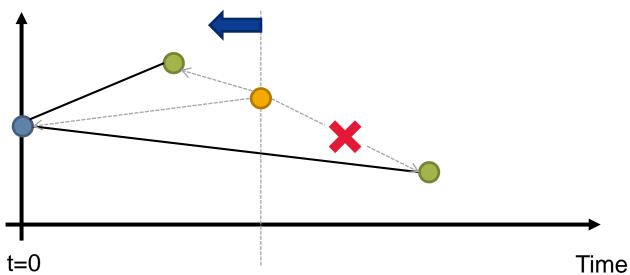
Spatio-temporal RRT\*





Spatio-temporal RRT\*

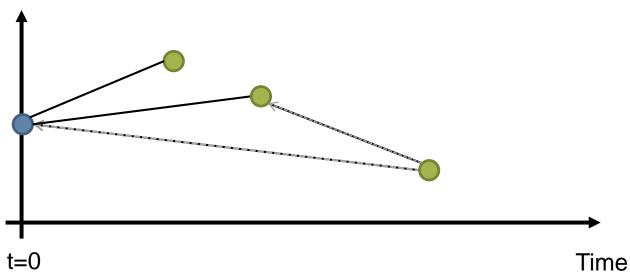
Mast configuration



 Samples must be connected in chronological order

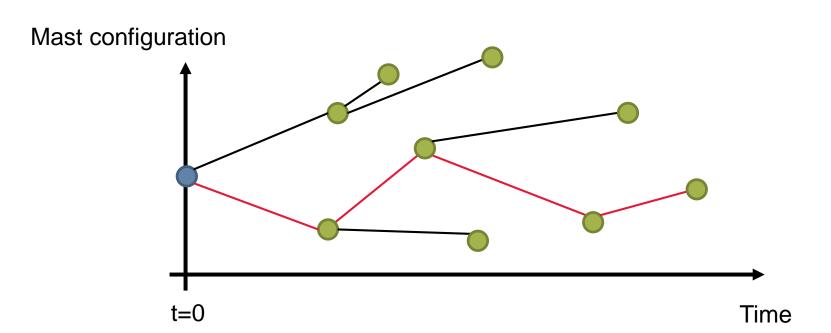
Spatio-temporal RRT\*

Mast configuration



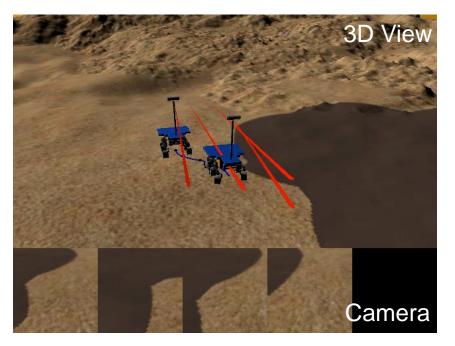
 Rewire connections to always keep optimal tree

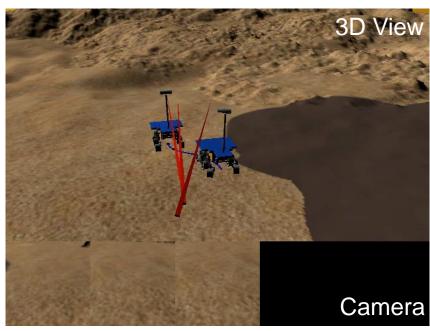
Spatio-temporal RRT\*



- Continue to grow the tree until it converges, or reaches max iteration
- A path with minimum leaf cost is the optimal mast trajectory

#### **Simulation**





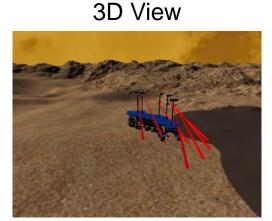
Fixed mast (baseline)

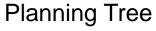
Proposed

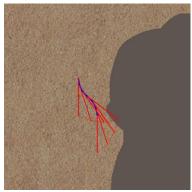
- JPL's DARTS/ROAMS Simulator
  - Used by past flight missions
  - Two regions: feature-rich (light) and feature-poor (dark)
- Performed spatio-temporal RRT\* in receding horizon manner (5 m horizon)

#### **Tree Evolution**

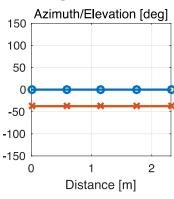
Fixed timing and direction (baseline)



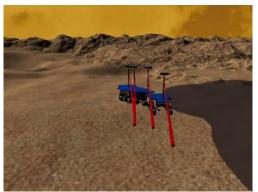




Angle Profile

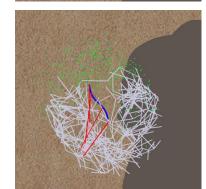


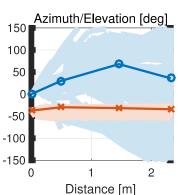
Iteration 120



Azimuth/Elevation [deg] 150 100 50 -50 -100 -150 2 0 Distance [m]

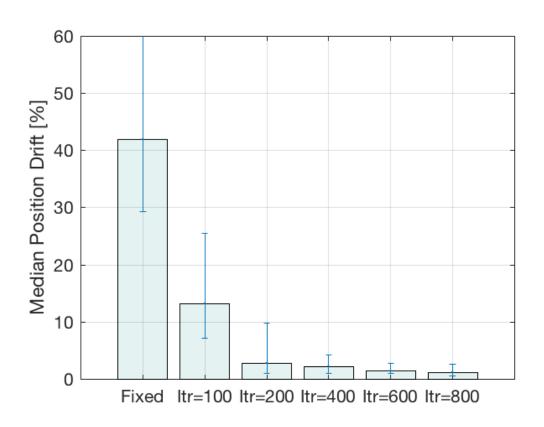
Iteration 800





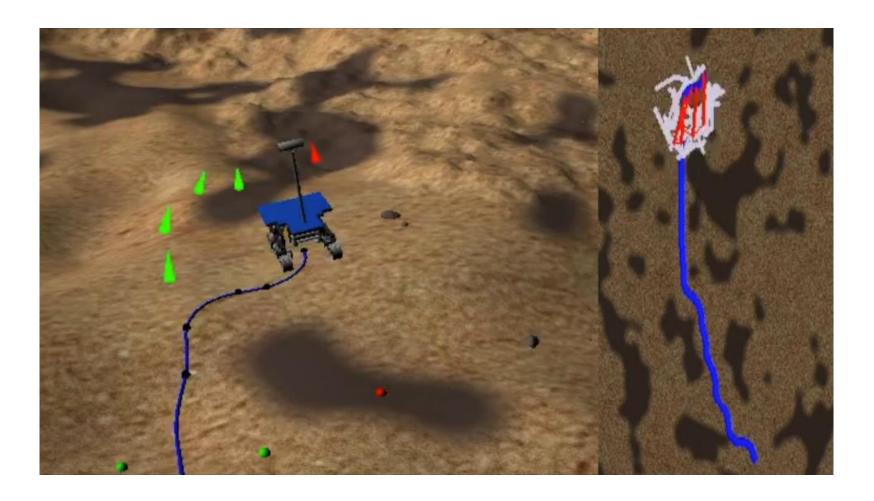
#### **Position Estimation Error**

Median error for 10 different maps



## **Receding Horizon Planning**

Replan every step within 5m horizon



#### Contributions:

- Proposed a method to actively steer visual sensors to improve autonomous navigation performance
- Developed an online search algorithm of mast trajectories using predicted future camera views

#### **Acknowledgement**:

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#### **Data Flow**

